

ESE3090 Final Project:

Empirical Banzhaf Indices

Mark Schulist

1) What is Power?

1.1) Naive Approach

The notion of power in a status quo election is much more complex than first meets the eye. In a status quo election, each voter is assigned a *weight*, and the item being voted on passes if the sum of the weights of the voters voting *yes* passes a threshold (*quota*).

On first thought, power seems like it must be proportional to the weight of each voter. For example, in the US electoral college, it seems reasonable that California's *power* should be $\frac{55}{538} \approx 0.10$. If this were true, then what would be the interpretation of this number? From our formulation of power, it's not entirely clear what the significance of this number corresponds to.

This naive approach to power breaks down with the following example. Imagine we have three voters and their corresponding weights $a = 49, b = 50, c = 2$ with quota $q = 51$. If we go through all possible ways that these three voters can win (ie the *winning coalitions*), we will find that voter b 's 50 votes are no more useful than voter a 's 49 votes. In other words, it is clear the a and b should have the same amount of power despite having different weights.

1.2) Banzhaf Power

From the above example, we can see that a more natural way of viewing power is as the *power to prevent*. Essentially, this boils down to "in how many winning coalitions is a voter pivotal?" Being pivotal means that if you take the voter out, that coalition would no longer win.

By reforming our notion of power to be *how often is a voter pivotal* makes the interpretation of power easier, but makes the math much more challenging. For one, the number of total coalitions (which is on the same order as the number of winning coalitions) is 2^m where m is the number of voters. This exponential growth means that computing power indices for large collections of voters (like the 51) is very difficult (we ignore generating functions as they will not help with the next step).

1.3) But wait, is this the entire power picture?

From a purely theoretical point of view, yes. But, as is often the case in applied math, the real world is full of experiences not available in our theoretical models. Remember, our goal is to compute the *power* in an election. Should power include prior information we have about an election? Does California really have as much power as a normal the normal Banzhaf indices say (about 11.4)? While California might be large, it seems unreasonable to say that it has a large amount of power considering its vote is pretty much guaranteed to be blue.

2) Empirical Power

2.1) Adopting a probabilistic mindset

In traditional Banzhaf indices, the probability of any given coalition *existing* is equal. This means that there is an equal chance of Texas and California voting the same as not voting the same. Of course, in reality there is a very unequal chance of this occurring.

To get a power index that relies on past election data, we need to *update* the probabilities of coalitions existing. In practice this looks like a lot of statistics, which has thankfully been done by the authors of [1].

$$\hat{p}_{i_1 \dots i_m} = \frac{1}{2} \frac{x_{i_1 \dots i_m} + x_{i'_1 \dots i'_m} + 1}{n + 2^{m-1}} \quad (1)$$

Equation 1 shows the posterior probability that a given coalition will exist. Each coalition is indexed by $i_1 \dots i_m$, where $i_j = 1$ if voter j is in the coalition and 0 otherwise.

The $x_{i_1 \dots i_m}$ (and its complement) represent our past data points. With n prior elections, the sum of all of the $x_{i_1 \dots i_m} = n$ because we have n prior data points. Clearly, if m is large and n is small, these prior data points will have a minuscule impact on the probabilities that a given coalition will occur as (most of the time) $x_{i_1 \dots i_m}$ will equal 0.

In the case where $n = 0$, where n is the number of past elections, all of the probabilities reduce to:

$$\hat{p}_{i_1 \dots i_m} = \frac{1}{2^m} \quad (2)$$

Equation 2 makes sense because without prior data (ie if we did not know that CA and TX will vote differently) every coalition should be equally likely to occur.

2.2) Probabilities to Power

Now that we have the probabilities of each coalition occurring, we need to transform them into the same notion of *power to prevent*. We want to calculate the conditional probability that a voter is pivotal given that they are in a winning coalition

$$P(v_i \text{ pivotal} \mid \text{winning coalition}) \quad (3)$$

To do this, we need to take the following sums

$$\frac{\sum_{S_1} p_{i_1 \dots i_m}}{\sum_{S_2} p_{i_1 \dots i_m}} \quad (4)$$

where S_2 is the set of all indices of winning coalitions and $S_1 \subseteq S_2$ such that the voter v of interest is pivotal in each coalition in S_1 .

$$S_1(v) = \{c \in S_2 \mid \text{voter } v \text{ pivotal in } c\} \quad (5)$$

Equation 5 shows that S_1 depends on the voter, so we will need to calculate S_1 for each voter in our election.

3) US Electoral College

3.1) Reality strikes again

It would be very nice if we could apply this probabilistic technique to get the “empirical” power for each US state. Unfortunately, there does not exist a nice generating function to obtain these indices for large number of voters, unlike for the traditional Banzhaf index.

To “solve” (more like work around) this problem, I decided to group together “similar” states (they have voted the same since 2000) to reduce the number of voters. Of course, grouping states can completely ruin the indices due to the discreteness of the problem. Additionally, to collapse the 51 states (DC included) into 20 while ensuring the each group had voted the same since 2000 meant that many of the psuedo-states (groups of states) are quite large while others are quite small.

3.2) Even sadder reality strikes again...

After building my psuedo-states and much debugging, I eventually was able to compute the empirical power indices for my 20 psuedo-states. Rather unfortunately, just 6 past elections was not enough to cause any significant change to the power indices.

Looking at Equation 1, we can see that the probabilities only change (a significant amount, if we ignore the $n \ll 2^{m-1}$) in a *very, very* small percentage of the coalitions. Sadly, this means that Equation 1 essentially just reduces to Equation 2 if we ignore minuscule differences.

Table 2 in [1] shows the empirical banzhaf power for each provinces in the 10 Canadian provinces. With only 10 provinces and 17 previous elections (half the number of voters and about triple the number of prior data elections compared the this paper), the probabilities do not change more than 1.2%. It is no wonder that Table 1 shows such little change...

States in Group	Empirical Index	Equally Likely Index
iowa	0.015229772418086636	0.01522994800373543
arizona	0.027973129808542167	0.027973452313699267
virginia	0.0329498028545735	0.032950182736317835
pennsylvania	0.050961900589514726	0.050962488134629204
missouri	0.02541369795629748	0.025413990953495433
nevada, colorado	0.038143604879061044	0.03814212312315775
georgia	0.04073762364822795	0.04073617179904001
indiana	0.027973129808542167	0.027973452313699267
alabama, mississippi, louisiana, oklahoma, alaska, west virginia, arkansas, south carolina, tennessee, wyoming, idaho, north dakota, south dakota, ne- braska, montana	0.1045158984528118	0.10451133887498991
michigan	0.04073570215284338	0.04073617179904001
utah	0.015229772418086636	0.01522994800373543
california, connecticut, delaware, hawaii, illinois, ore- gon, maryland, washington, massachusetts, new jer- sey, new york, district of columbia, rhode island, vermont, maine	0.9029087597131591	0.9029172479257218
florida	0.07376620781356874	0.07376321523686546
new hampshire	0.010218512455133071	0.010218630265438433
wisconsin	0.02541369795629748	0.025413990953495433
texas, kansas, kentucky	0.1017335731359572	0.10172898148041397
ohio	0.045714296694259283	0.04571290222165858
new mexico	0.01267802654738022	0.012678172713682358
minnesota	0.02541369795629748	0.025413990953495433
north carolina	0.03814360487906103	0.03814212312315775

Table 1: Banzhaf Indices Computed Using [1] for Election Data from 2000-2020

Table 1 does show something interesting (and somewhat unrelated): there are two very large blocks of US states that have always voted the same since 2000 (until 2020). While I was attempting to group states under the constraint of ensuring that they have voted the same in all previous elections, I was forced to create the 2 large psuedo-states in order to keep the number of “voters” to 20. In some sense, this grouping alone represents what it means to be a swing state (where any state outside of the 2 large groups could be considered a swing state).

3.3) Effect of grouping states

While my idea of using prior election data may not be feasible (at least using the approach in [1]), we can still compare the effect that grouping states had on their power.

We know the traditional power indices for each state, so we can see if the sum of the indices for each state in each psuedo-state aligns with the power for each psuedo-state.

States in Group	Sum of Banzhaf	Scaled Grouped Index
Alabama, Mississippi, Louisiana, Oklahoma, Alaska, West Virginia, Arkansas, South Carolina, Tennessee, Wyoming, Idaho, North Dakota, South Dakota, Ne- braska, Montana	0.159	0.06311648525077171
Pennsylvania	0.037000000000000005	0.030777264604172957
Utah	0.011000000000000001	0.009197669830807437
New Hampshire	0.006999999999999999	0.006171234943254347
California, Connecticut, Delaware, Hawaii, Illinois, Oregon, Maryland, Washington, Massachusetts, New Jersey, New York, District of Columbia, Rhode Island, Vermont, Maine	0.35800000000000001	0.5452897625733981
Wisconsin	0.018000000000000002	0.015348016803212109
North Carolina	0.027000000000000003	0.023034790075892965
New Mexico	0.009000000000000001	0.0076566017592313235
Virginia	0.024	0.019899273562791565
Texas, Kansas, Kentucky	0.098	0.06143616404019775
Georgia	0.028999999999999998	0.024601387889618683
Ohio	0.033	0.0276069348063221
Missouri	0.018000000000000002	0.015348016803212109
Michigan	0.028999999999999998	0.024601387889618683
Nevada, Colorado	0.027000000000000003	0.023034790075892965
Arizona	0.02	0.01689372664608815
Florida	0.054000000000000006	0.04454707916540952
Minnesota	0.018000000000000002	0.015348016803212109
Iowa	0.011000000000000001	0.009197669830807437
Indiana	0.02	0.01689372664608815

Table 2: Comparison of the sum of individual Banzhaf indices and the index from the entire group

The reason that the grouped index and the empirical Banzhaf indices are so different is that the sum of the empirical Banzhaf indices do not equal one (necessarily), so I have normalized them here so that the comparison with the sum of the traditional Banzhaf indices are comparable.

Most of the states stay somewhat constant, with the large blue group (with CA and NY) becoming much more powerful and the large red block (with Alabama and Mississippi) becoming far less powerful.

While I do not have a conclusive answer to why this is the case, I believe that because the large blue group has vastly more votes than any other group, they *almost* have veto power in the election. Their (near) veto power allows them to have very high empirical Banzhaf score (around 0.9), which means that they are necessary in 90% of winning coalitions.

Attempting to group states changes the dynamics of this discrete system in unexpected ways, especially under the constraint that all states in a single group must have voted the same in past elections.

Bibliography

[1] A. Heard and T. Swartz, “Empirical Banzhaf Indices,” *Springer Nature*, vol. 97, no. 4, pp. 701–707, 1998.

Code is available at <https://gist.github.com/mschulist/fa2d7a8e6beaf7ed0bf7bad4a85e0ad2>